

# Purpose

The purpose of the section is to help you learn how to research, select, and develop appropriate algorithms to become a Successful Artificial Intelligence (AI) Engineer

At the end of this lecture, you will learn the following

 How does Davies-Bouldin Index compute the average similarity between each cluster and its most similar cluster





## How to determine type of output and evaluation metrices?

# Understand the problem

Domain

Objectives

Constraints

Define the problem as a

Supervised

Unsupervised

Reinforcement learning task.

Determine

type of output (e.g., classification, regression, clustering)

**Evaluation metrics** 



Silhouette Score

Clustering-Evaluation Metrics

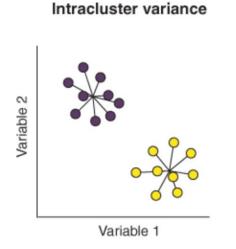
Adjusted Rand Index (ARI) and Adjusted Mutual Information (AMI) Davies-Bouldin Index

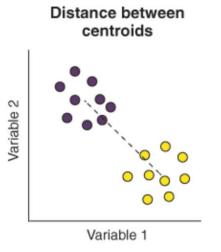




## Calculating the Davies-Bouldin Index

- •For each cluster *i*:
  - •Compute the centroid *Ci*, which represents the center of the cluster.
  - •For each other cluster j (where  $j \neq i$ ):
    - Calculate the average distance between each point in cluster i and the centroid
       Ci. This is denoted as avgi.
    - Calculate the average distance between each point in cluster j and the centroid
       Cj. This is denoted as avgj.
    - Compute the distance between the centroids *Ci* and *Cj*. This is denoted as *dij*.
  - Calculate the similarity between cluster *i* and its most similar cluster *j* as Rij=(avgi+avgj)/dij
- The Davies-Bouldin Index for the clustering is the average of the similarity values *Rij* across all clusters











## Interpreting the Davies-Bouldin Index

# Low Davies-Bouldin Index

- Clusters are well-separated and distinct from each other
- Each cluster is more similar to its own cluster than to others

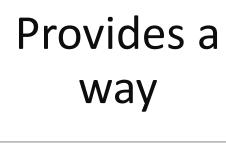
High
DaviesBouldin
Index

- Clusters are less wellseparated and may have significant overlap
- The boundaries between clusters are less clear



Enrichmentors

## Using the Davies-Bouldin Index for Cluster Evaluation



- Quantitatively evaluate
- Quality of a clustering algorithm

#### Compare

- Different clustering algorithms or parameter settings
- Select the optimal number of clusters.

### Complements

- Other clustering evaluation metrics, such as the Silhouette Score
- Provides additional insights into the clustering quality







Silhouette Score

Clustering-Evaluation Metrics

Adjusted Rand Index (ARI) and Adjusted Mutual Information (AMI)

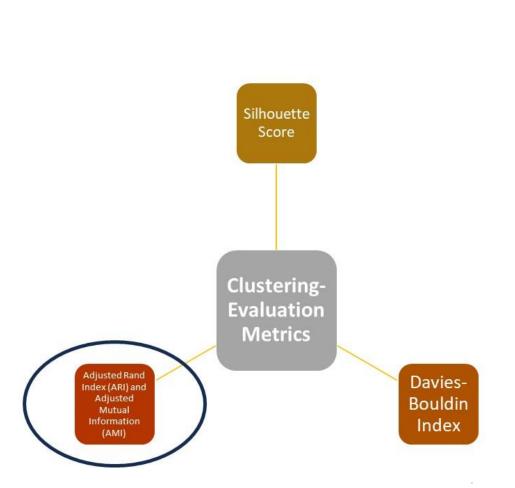
Davies-Bouldin Index





# What is next?

Adjusted Rand Index (ARI) and Adjusted Mutual Information (AMI) Measures the agreement between true labels and cluster assignments







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